

# Data-analytical Stability in Second-Level fMRI inference

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## problem

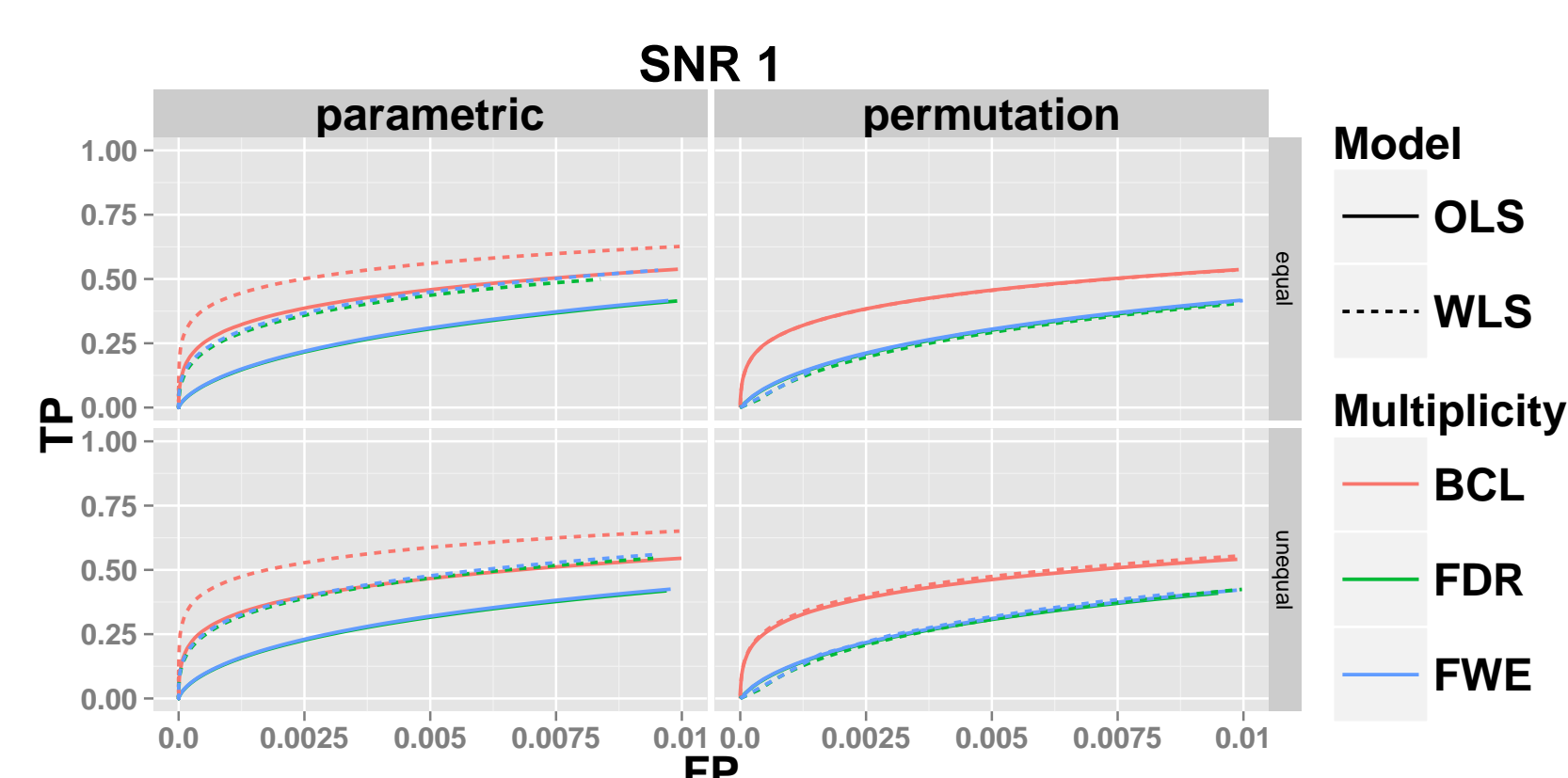
- ◇ **fMRI** plays an important role to localize brain regions
- ◇ 2-level **mass-univariate** GLM-based approach still popular for multi-subject analysis
- ◇ in voxel-based analysis: 1) continued use of **different models** to pool information over subjects; 2) **various multiple testing** procedures to control false positives and 3) **distinct inference procedures**
- ◇ choices in each of three phases impact reproducibility
- ◇ today no golden standard for a voxel-based approach
- ◇ **data-analytical stability** has the ability to further inform these choices, while it is also a proxy for **reproducibility** and intrinsically allows for an evaluation on real data

## model - inference - multiple testing

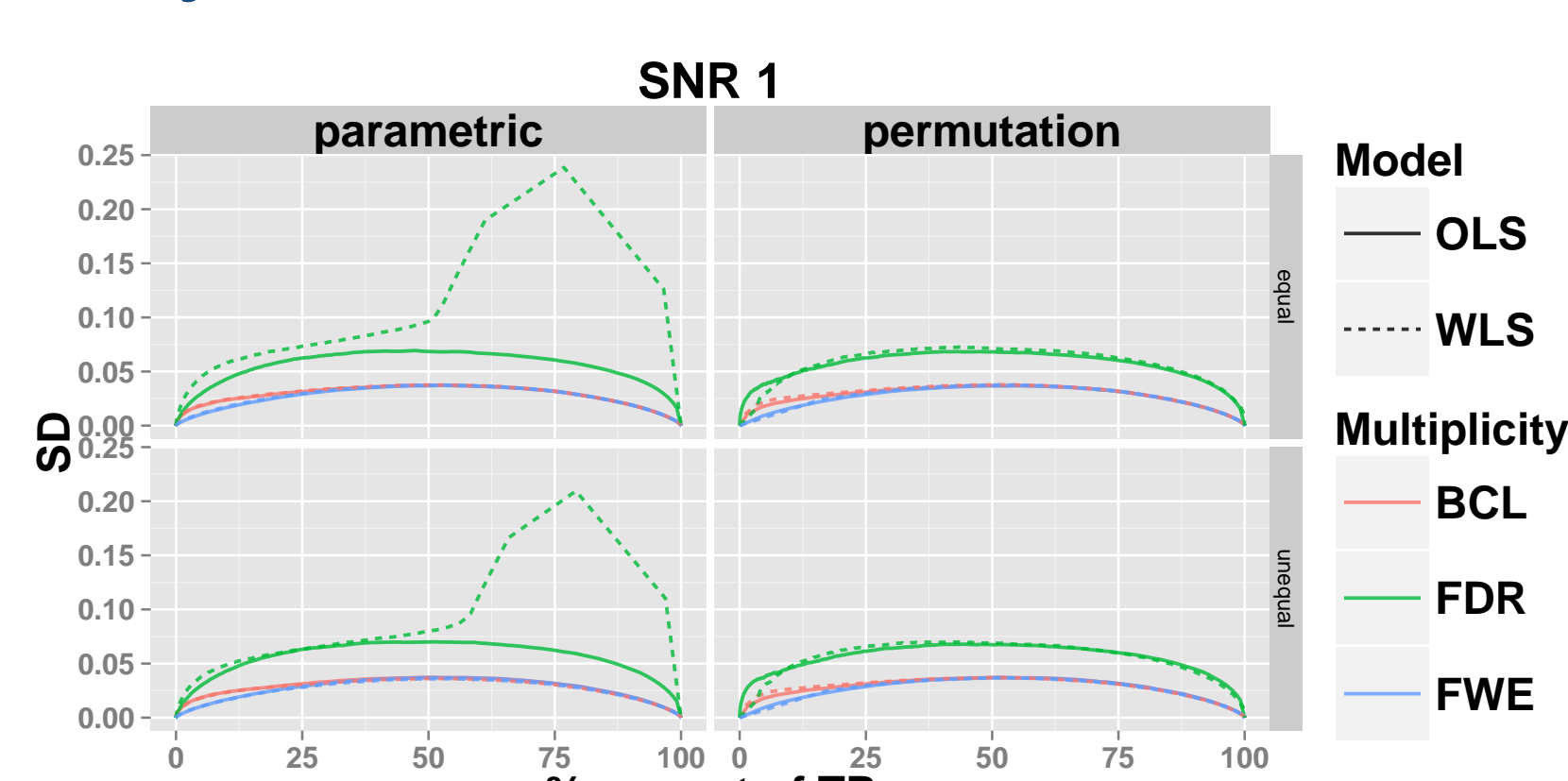
- ◇ from **1st level model** the contrast image  $c\hat{\beta}$  from voxel-wise model  $Y_v = X\beta_v + \epsilon_v$  is fed into
- ◇ **the 2nd level GLM**[1]: yet no consensus on how to pool information over subjects [2, 3]
  - **OLS** ignores subject-specific variability (only uses  $c\hat{\beta}$ )
  - **WLS** accounts for subject specific variability (also includes the use of  $var(c\hat{\beta})$ ) via down-weighting
- ◇ **account for multiple testing**: also no standard but choice between theory-driven [4] and practical approaches [5] for **inference** based on parametrical assumptions or on permutations [3]
  - **FWE**: control probability of at least 1 False Positive (FP) (via RFT (parametrical inference) or *max* statistic (permutation))
  - **FDR**: control proportion of FPs over all selected voxels (via *p*-values that are computed via parametrical inference or permutation)
  - **BCL**: control FPs by setting a voxel-wise (**B**onferroni-like) threshold and specify a minimal **CL**uster size (here 10)

## simulation study

### balance FP and TP



### stability



- ◇ WLS outperforms OLS
- ◇ BCL outperforms FDR and FWE
- ◇ permutation no diff. between OLS and WLS
- ◇ for higher SNR, smaller differences
- ◇ FDR is always more variable
- ◇ BCL comparable with FWE
- ◇ permutation  $\approx$  parametrical inference
- ◇ high SNR, FDR + permutation: very instable

## goals

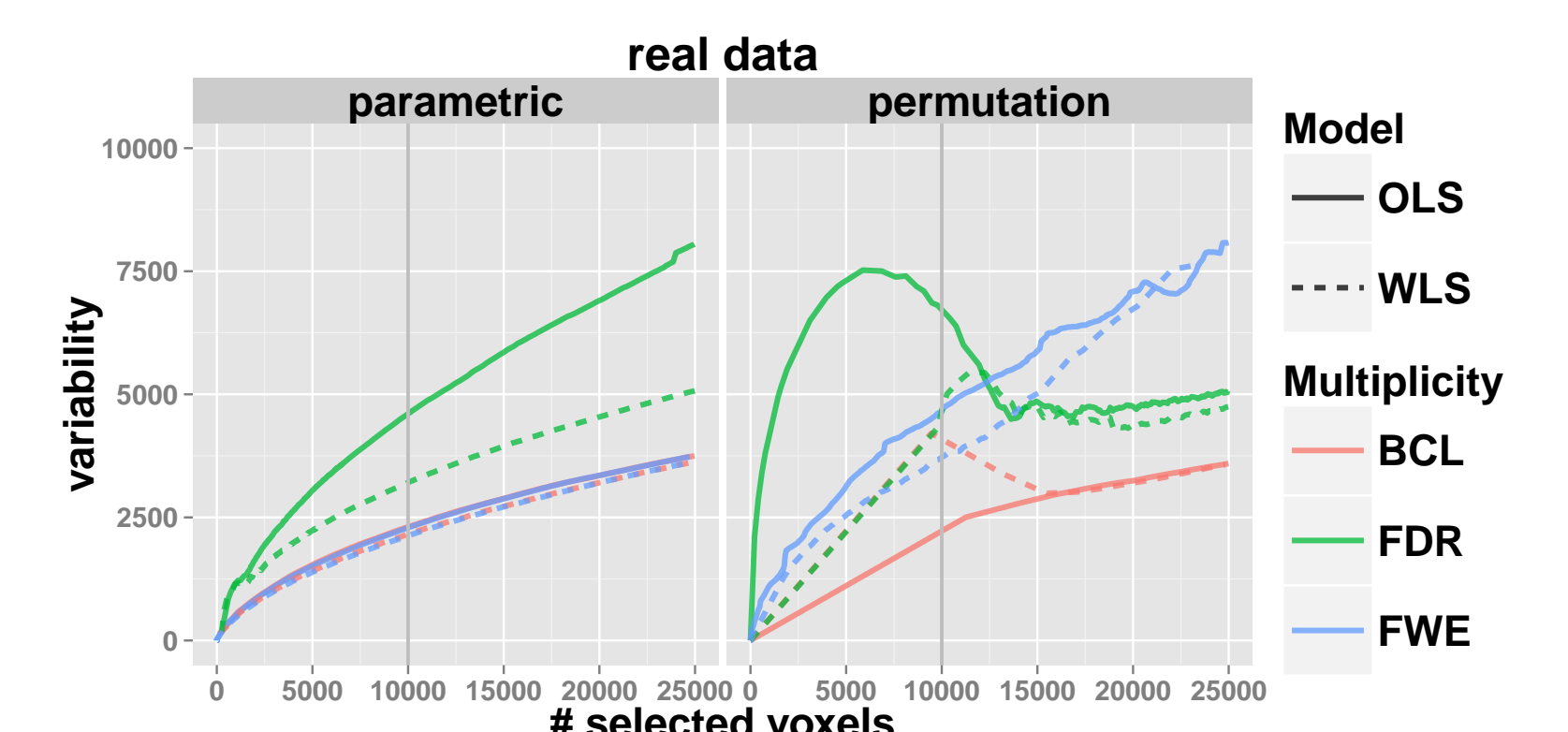
- ◇ **comparison** of combinations of methods for analysis of multi-subject fMRI
- ◇ to achieve a **golden standard** for analysis
- ◇ illustrate how to evaluate based on both **simulation and real data**

## methods

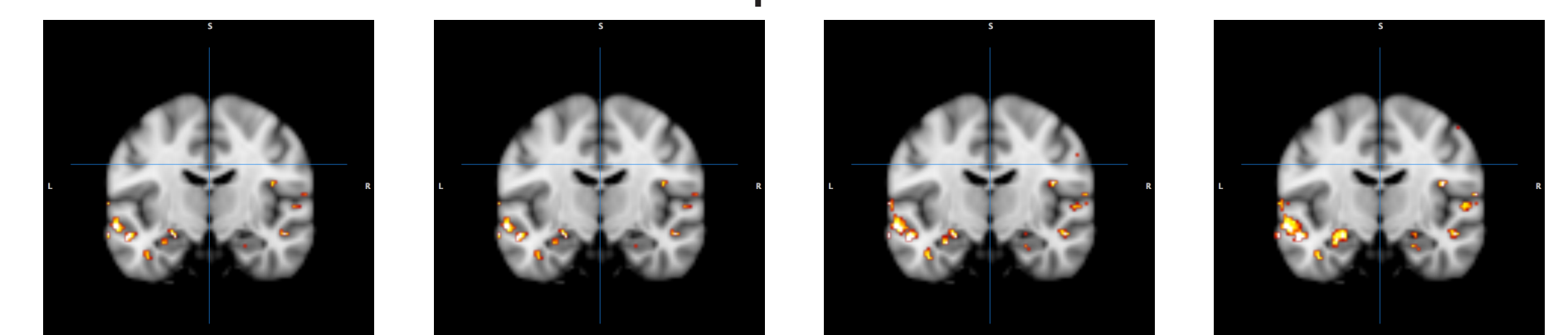
- ◇ **simulation** 2 clusters in  $45 \times 45 \times 45$  volume; no first-level noise; smoothing applied to comply RFT assumptions; 2 SNR settings
- ◇ **real data**: 15 subjects from HCP data
- ◇ **criteria**
  - **type I&II error rates** Receiver-Operator-Curve False and True Positives (**FP,TP**) versus False and True Negatives (**FN,TN**)
  - **stability** variability on the number of selected voxels and *re-selection rate*: at a fixed threshold compute the # a voxel is selected over K bootstrap samples
- ◇ **procedures** FSL-based, in-house R-based

## real data

### stability



### re-selection rates for parametrical inference



- ◇ FDR: more variable + has lower re-selection rates
- ◇ WLS is found to be less variable

## discussion and conclusions

- ◇ **model** WLS is slightly more stable than OLS, in ROC WLS outperforms OLS
- ◇ **multiple testing correction** FWE and BCL are almost always more stable than FDR, in ROC BCL outperforms FDR/FWE
- ◇ **inference** both for FP/TP balance and stability we find no large differences between permutation-based and parametrical inference
  - with respect to FP/TP balance, BCL offers a good alternative, but lacks statistical justification
  - with respect to stability, FDR is outperformed by the other procedures

## references

- [1] Beckmann, et al. (2003) General multi-level linear modeling for group analysis in fMRI, *NeuroImage*
- [2] Mumford and Nichols (2009). Simple group fMRI modeling and inference, *NeuroImage*
- [3] Thirion, et al. (2007). Analysis of a large fMRI cohort: . . . *NeuroImage*
- [4] Bennett et al. (2009). The principled control of false positives in neuroimaging. *SCAN*
- [5] Lieberman et al. (2009) Type I and Type II error concerns in fMRI research: . . . *SCAN*

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## inter@ctive



<http://tinyurl.com/j9qsjjd>